Control of 2D Minimally Persistent Formations with Three Co-Leaders in a Cycle

Regular Paper

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Received 10 Sep 2012; Accepted 18 Oct 2012

DOI: 10.5772/54494

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Abstract This paper mainly solves the problem of 2D minimally persistent formation control in which three co-leaders are in a cycle, which raises a great challenge in the area of persistent formation control. First, a novel control law is proposed for this problem. The fundamental moving principles of the agents are well-designed based on the property of persistence, and the non-square rigidity matrix is converted into the square one for the design of the control law. Then, the method of leading principal minor is utilized to prove that the formation with the above control law can be stabilized. Finally, simulation results show that the proposed controllers are able to stabilize the group formation to a rigid shape, while keeping the distances between the agents to the desired value.

Keywords Three-co-leader, Minimally Persistent Formation, Rigidity Matrix

1. Introduction

In recent years, many research papers [1-10] have indicated different ways of solving the problem of fulfilling tasks using a formation. This paper however proposes a minimally persistent formation controller to maintain the distance of the agents continuously based on the concept of rigid control [11, 12]. Furthermore, different from the previous work, what we are dealing with is formation control based on a directed graph, rather than undirected. This type of the rigid formation has been termed as the persistent formation [13-23].

The underlying graph of the directed formations may be either cyclic or acyclic. Formation control with an acyclic graph is easy dealt with because of its particular structure, where the follower agents cannot influence the leader agents. However, formation control with a cyclic graph has well-known difficulties, since the control of the leader agents may be affected by the follower [18]. In the area of distance maintenance control with a cyclic graph, some authors have made some contributions [15, 23-25]. Baillieul [19] proposed a method of formation control considering the distance measurements in a cyclic structure, but he did not research how to maintain the distance of all the agents. Hendrix [15] then discussed the possibility of keeping the distance between every pair of agents constant in a cycle graph and
raised the concept of minimally persistent formation, but the specific control of the whole formation was not analysed. Furthermore, Yu [17] researched the special control law in a minimally persistent formation with a leader-first-follower variety in the plane and tried to design the control law using a rigidity matrix. However, he only considered the one-leader situation in which the followers are in a cycle. Moreover, in the research of leaders in a cycle, Anderson [25] considered a control law for a formation with only three agents named three co-leaders, but he did not explain the topology relationship of the followers.

To the best of our knowledge, all the aforementioned papers share the following common drawbacks: one-leader model for minimally persistent formation cannot complete the complex mission very soon and it is difficult for one leader to find out the goal in a complicated environment as soon as possible. Consequently, the one-leader model is an important factor limiting the performance of the whole formation. In addition, all the algorithms based on the one-leader model seem to have poor scalability and lack adaptability and flexibility to both tasks and environment.

To overcome the aforementioned drawbacks, a multi-leader architecture is necessary. Yu [17] mentioned that only the minimally persistent formation systems with three leaders can be stabilized. Therefore, the problem of minimally persistent formation control with three co-leaders is discussed in this paper. In the three co-leader model, the three co-leaders are equivalent. In addition, the non-square rigidity matrix, which indicates the distance of all the agents, is introduced in the design of the control law to deal with the timeliness problem.

The primary contribution of this paper is the proposing of a novel control law using a non-square rigidity matrix for minimally persistent formations, in particular, under the condition that the three co-leaders are in the cycle. Firstly, since the leaders and followers have different types of the moving principle, we devise different control laws for them respectively. Secondly, during the course of designing the control laws, the novel method of converting the non-square rigidity matrix into the square one is discussed in detail. Thirdly, it is necessary to consider the situation where the followers are in a cycle and when they are not. Finally, we do the simulations in which the followers are in the acyclic and cyclic graphs to prove the efficiency of the proposed control law.

2. Preliminaries

2.1 Graph rigidity

Before introducing persistent formations in a directed graph, it is necessary to know the underlying rigid formations in an undirected graph. Let \( G=(V,E) \) be an undirected graph with \( n \) vertices; denote the composite vector \( p=(p_1,p_2,\ldots,p_n) \in \mathbb{R}^{2n} \) and a pair \((G,p)\). \( p_i \ (i \in [1,2,\ldots,n]) \) represents the coordination of the \( i \)th vertex in the place.

Denote the constant parameter \( d_{ij} = ||p_i - p_j|| \) as the Euclidean distances between pairs of points \((p_i,p_j)\). In addition, the introduction of the rigidity matrix in [11] claims that:

\[
\{(p_i - p_j), (p_i - p_j)\} = d_{ij}^2 \quad i,j \in [1,2,\ldots,n] \quad (1)
\]

Assume that the trajectory is smooth, then we could get the following expressions from (1):

\[
2\{(p_i - p_j), (p_i - p_j)\} = 0, \quad i,j \in [1,2,\ldots,n], \quad t \geq 0 \quad (2)
\]

where \( p_i \) is the velocity at point \( p_i \). Then, we obtain the following homogeneous equation by combining (2) at different points:

\[
\sum_{i,j=1}^{2n} R_{ij} = 0 \quad (3)
\]

where \( R = \text{column}\{p_1, p_2, \ldots, p_n\} \) and \( R \) is called the rigidity matrix with dimension \( n \times 2n \ [n = n(n-1)/2] \) in the plane.

In addition, the rigidity function is described as follows for another definition of rigidity matrix:

\[
\hat{\gamma}_C(p) = ||p_i^{in} - p_i^{out}||^2 + ||p_j^{in} - p_j^{out}||^2 + \ldots + ||p_k^{in} - p_k^{out}||^2 \quad (4)
\]

where the \( i \)th component \( ||p_i^{in} - p_i^{out}||^2 \) corresponds to the edge \( e_k \) of \( E \ (k \in [1,2,\ldots,m]) \). \( E = \{e_1,e_2,\ldots,e_m\} \).

**Definition 2.1** [12, 26, 27] A framework \((G,p)\) is infinitesimally rigid in the plane if \( \text{dim}(\ker J_\gamma(p)) = 3 \), or if

\[
\text{rank}_{\hat{\gamma}_C}(p) = 2n - 3 \quad (6)
\]

where \( J_\gamma(p) \) is the Jacobian matrix of \( \hat{\gamma}_C(p) \) and \( \text{rank}_{\hat{\gamma}_C} \) has the same form as the rigidity matrix \( R \).

**Definition 2.2** [12] Point \( p \) is a regular point of graph \( G \) with \( n \) vertices if

\[
\text{rank}_{\hat{\gamma}_C}(p) = \max_{q \in \mathbb{R}^{2n}} \{\text{rank}_{\hat{\gamma}_C}(q)\} \quad (7)
\]

**Lemma 2.1** [26, 27] A framework \((G,p)\) is infinitesimally rigid if and only if \((G,p)\) is rigid and \( p \) is a regular point.
Lemma 2.1 tells us that some frameworks are rigid but not infinitesimally rigid. However, if the framework is infinitesimally rigid, then it is sure to be rigid. Fig. 1 illustrates these properties with two examples. It is easy to compute that rank $b_k(q) = 2n - 3$ in Fig. 1 (a) and rank $b_k(q) = 2n - 3$ in Fig. 1 (b), so Fig. 1 (a) is rigid and infinitesimally rigid; Fig. 1 (b) is rigid but not infinitesimally rigid as $p$ is not a regular point. In general, the rigid graphs which fail to be infinitesimally rigid almost have parallel or collinear edges. In this paper, “rigid” almost always means “infinitesimally rigid”.

**Lemma 2.2** [28] Consider a formation $F$ in $R^2$, with agents in generic positions and with defined agent pairs having the agent-agent distances maintained. Let $G = (V, E)$ be the undirected graph. Then $F$ is rigid if and only if there is a subset $E' \subseteq E$ satisfying the following two conditions:

1. $|E'| = 2|V| - 3$.
2. For all $E'' \subseteq E'$, $E'' \neq \emptyset$.

$$|E''| \ll 2|V(E'')| - 3,$$

where $|V(E'')|$ is the number of vertices which are the end vertices of the edges in $E''$.

**2.2 Persistent and minimally persistent graph**

Rigidity is the property of the undirected graph, and persistence is the corresponding property of the directed graph.

**Lemma 2.3** [15] A representation is persistent if and only if it is rigid and constraint consistent. A graph is generically persistent if and only if it is generically rigid and generically constraint consistent.

From Lemma 2.3, we know that rigidity and constraint consistency are the crucial factors for the persistence of a directed graph. However, in the plane, the rigidity of a directed graph is actually the rigidity of the corresponding underlying undirected graph, and the constraint consistency means that every agent should satisfy all their own distance constraints.

As a result, the graph in Fig. 2 (a) is persistent and the graph in Fig. 2 (b) is not persistent as node 2 is not constraint consistent (one node is able to satisfy two distance constraints at most in the plane). The arrows in the figures indicate the “leading” relationship, instead of the direction of the information flow which is commonly used in graph persistent theory [14]. As a result, the arrows indicating node 4 in Fig. 2 (a) mean that node 4 is the “leader” of node 1, 2 and 3; node 1, 2 and 3 can only know the information of node 4.

The minimal rigid graph requires that the graph has the least edges to satisfy the rigid conditions, and if one edge is removed, the graph will not be rigid. A comparative relationship between the rigidity and persistence is that the conditions of the minimally persistent graph are more complicated than that of the rigidity graph.

**Lemma 2.4** [15] Consider a directed graph with more than one vertex. Then it is minimally persistent if and only if the underlying undirected graph is minimally rigid and no vertex has more than two outgoing edges.

Lemma 2.3 and Lemma 2.4 tell us that the number of the vertices in a minimally persistent graph is always $2n-3$ ($n$ represents the number of nodes).

**3. Formation control laws**

An example of minimally persistent formation with three co-leaders is shown in Fig. 3 in which agent 4, 5 and 6 are the leaders in a cycle and each of them has one Degree of Freedom (DF). The DF of any of the followers 1, 2 and 3 is zero. All the followers are directly or indirectly controlled by the leaders to fulfil the distance constraint, i.e., the motion of agent 1 is constrained by agents 4 and 6; while the motion of agent 2 is constrained by agent 4 and agent 1.
In this paper, the first order kinematic model is adopted for every agent:

$$u_i = p_i \dot{}$$  

(8)

### 3.1 Control laws for followers

For $n$ agents, the index of all the followers is defined from 1 to $n-3$. When one agent $p_i$ has to maintain the constant distance ($d_{ij,0}$ and $d_{ik,0}$) from two leader neighbours ($p_j$ and $p_k$), it has two choices (shown in Fig. 4). Agent $p_i$ may choose $p_{i,o}$ or $p_{i,o}^\prime$ as its targeted position since both of them have the same distance ($d_{ij,0}$ and $d_{ik,0}$) to $p_j$ and $p_k$. However, the position $p_{i,o}$ is more feasible than $p_{i,o}^\prime$ due to a short moving distance. Therefore, $p_{i,o}$ is chosen as the target position to be reached for the follower agent $p_i$. All the followers in this paper are guided by such a principle.

![Figure 4. Moving principle of the followers.](image)

From (8), we can design the control law:

$$u_i = p_i = K_i(p_{i,o} - p_i) = K_i f(p_j - p_i)p_k - p_i, d_{ij,0}d_{ik,0})$$

where $K_i$ is the gain here.

Suppose that the position of agent $p_i$ can be expressed as $p_i(t) = p_i + \delta p_i(t)$, where $p_i(t) = (x_i, y_i)$ is the desired position which satisfies the dynamic constraint on the distance for the $i$th agent, and $\delta p_i(t)$ means a small variable. From Fig.4, by using the cosine law to the triangle $p_ip_jp_{i,o}$ and neglecting the non-linear second order terms, a linear equation can be obtained as follows:

$$2(x - x_i) - y_i - (y - y_i) \rangle (p_{i,o} - p_i) \approx d_{ij,0}^2 - d_{ik,0}^2$$

(10)

where $d_{ij,0}$ represents the distance between $p_i$ and $p_j$, and $d_{ik,0}$ means the distance between $p_i$ and $p_k$.

Then,

$$p_{i,o} - p_i \approx \frac{1}{2} [x - x_i - y_i - y_i] \begin{bmatrix} d_{ij,0}^2 - d_{ik,0}^2 \\ d_{ij,0}^2 - d_{ik,0}^2 \end{bmatrix}$$

As a result, the control law of the followers is represented by:

$$\frac{1}{2}[d_{ij,0}^2 - d_{ik,0}^2] \begin{bmatrix} \dot{p}_{ij} \\ \dot{p}_{ik} \end{bmatrix}\delta p_i - \begin{bmatrix} \dot{p}_{ij} \\ \dot{p}_{ik} \end{bmatrix}\delta p_i$$

(11)

where $R_{ij,ik}$ is the sub-matrix of the rigidity matrix $R$. (Please see [17] for more details).

So, for all the followers, the control law can be obtained as follows:

$$\dot{\delta p}(t) = KR_c^{-1}R\delta p(t)$$

(13)

where both $K$ and $R_c$ are $2 \times 2$ diagonal block matrices. $R_c$ partly represents the rigidity of the formation.

### 3.2 Control laws for co-leaders

There are three co-leaders indexed by $n-2$, $n-1$, and $n$, and they lead each other. The minimally persistent formation controllers for three co-leaders are proposed as:

$$p_{n-2} = k_{n-1} p_{n-3} - \frac{1}{1||p_{n-1} - p_{n-2}||}d_{(n-2,n-1),o}(p_{n-1} - p_{n-2})$$

$$p_{n-1} = k_{n-1} p_{n} - \frac{1}{1||p_{n-2} - p_{n-1}||}d_{(n-2,n),o}(p_{n-2} - p_{n-1})$$

$$p_n = k_{n} p_{n-2} - \frac{1}{1||p_{n-2} - p_n||}d_{(n-1,n-2),o}(p_{n-2} - p_n)$$

(14)

where $k_{n}, k_{n-1}$ and $k_{n-2}$ are the gains.

Then
Similarly, where

\[
\begin{bmatrix}
\delta x_{n-2} \\
\delta y_{n-2}
\end{bmatrix} = k_{n-2} I R^{-1}_{c,n-2} R((n-2),(n-1),0) \begin{bmatrix}
\delta x_{n-2} \\
\delta y_{n-1}
\end{bmatrix}
\]  

(15)

where

\[
R_{((n-2),(n-1),0)} = \begin{bmatrix}
\bar{x}_{n-2} - \bar{x}_{n-1} & \bar{y}_{n-2} - \bar{y}_{n-1} \\
\bar{x}_{n-1} - \bar{x}_{n-2} & \bar{y}_{n-1} - \bar{y}_{n-2}
\end{bmatrix}
\]

(16)

\[R_{c,n-2}\] is the sub-matrix of \(R_c\) corresponding to node \(n-2\). Refer to (10), the \(R_{c,n-2}\) becomes:

\[
R_{c,n-2} = \begin{bmatrix}
\bar{x}_{n-1} - \bar{x}_{n-2} & \bar{y}_{n-1} - \bar{y}_{n-2} \\
-(\bar{y}_{n-1} - \bar{y}_{n-2}) & -(\bar{x}_{n-1} - \bar{x}_{n-2})
\end{bmatrix}
\]

(17)

Each row of \(R_{c,n-2}\) is derived based on (10). As a result, the control law for \(p_{n-2}\) is:

\[
\delta \dot{p}_{n-2}(t) = k_{n-2} I R^{-1}_{c,n-2} R((n-2),(n-1),0) \delta \dot{p}_{n-2}(t)
\]  

(18)

Similarly, we get:

\[
\delta \dot{p}_{n-1}(t) = k_{n} I R^{-1}_{c,n-1} R((n-1),(n),0) \delta \dot{p}_{n-1}(t)
\]  

(19)

where

\[
R_{((n-1),(n),0)} = \begin{bmatrix}
\bar{x}_{n-1} - \bar{x}_{n} & \bar{y}_{n-1} - \bar{y}_{n} \\
\bar{x}_{n} - \bar{x}_{n-1} & \bar{y}_{n} - \bar{y}_{n-1}
\end{bmatrix}
\]

(20)

and

\[
\delta \dot{p}_{n}(t) = k_{n} I R^{-1}_{c,n} R((n-2),(n-1),0) \delta \dot{p}_{n}(t)
\]  

(21)

3.3 Control laws for the whole formation

Without loss of generality, any 2D linear motion can be equivalent to a two-step motion consisting of translation along the \(x\)-axis and \(y\)-axis. To simplify the analysis process, a two-step translation along the \(x\)-axis and \(y\)-axis are used to substitute the linear motion in this paper.

For the first step of the translation along the \(x\)-axis, the leader who firstly discovers the goal is defined as node \(n\). In order to realize the \(x\)-axis translation, let the values of \(\delta y_n, \delta \bar{x}_n, \delta \bar{y}_{n-1}\) and \(\delta \bar{y}_{n-2}\) all be zero, hence the values of \(\bar{y}_{n-2} - \bar{y}_{n-1}, \bar{y}_{n} - \bar{y}_{n-1}\) and \(\bar{y}_{n-2} - \bar{y}_{n}\) are also zero.

So we get that:

\[
\begin{bmatrix}
\delta \dot{x}_{n-2} \\
\delta \dot{y}_{n-2}
\end{bmatrix} = \begin{bmatrix}
-k_{n-2}(\delta \bar{x}_{n-2} - \delta \bar{x}_{n-1}) \\
0
\end{bmatrix}
\]

(22)

Likewise, for the second step of the translation along the \(y\)-axis, we arrive at the following conclusion:

\[
\begin{bmatrix}
\delta \dot{x}_{n-1} \\
\delta \dot{y}_{n-2}
\end{bmatrix} = \begin{bmatrix}
-k_{n-1}(\delta \bar{x}_{n-1} - \delta \bar{x}_{n}) \\
0
\end{bmatrix}
\]

(23)

\[
\begin{bmatrix}
\delta \dot{x}_{n} \\
\delta \dot{y}_{n}
\end{bmatrix} = \begin{bmatrix}
0 \\
-k_{n}(\delta \bar{y}_{n-2} - \delta \bar{y}_{n-1})
\end{bmatrix}
\]

(24)

\[
\begin{bmatrix}
\delta \dot{x}_{n-1} \\
\delta \dot{y}_{n}
\end{bmatrix} = \begin{bmatrix}
0 \\
-k_{n}(\delta \bar{y}_{n-1} - \delta \bar{y}_{n-2})
\end{bmatrix}
\]

(25)

Since the translation along the \(x\)-axis and \(y\)-axis is equivalent, we can only discuss the case of translation along the \(x\)-axis here.

With (11), (21), (22) and (23), we obtain a new control law:

\[
\begin{bmatrix}
\delta \dot{x}_{n-2} \\
\delta \dot{y}_{n-2}
\end{bmatrix} = \begin{bmatrix}
-k_{n-2}(\delta \bar{x}_{n-2} - \delta \bar{x}_{n-1}) \\
0
\end{bmatrix}
\]

(26)
The number of the rest control variable is only (2n-4) since $\delta y_{n-2}$, $\delta y_{n-1}$ and $\delta y_{n-2}$ are all zero, so both $\hat{R}$ and $R$ are $(2n-4) \times (2n-4)$. As $R$ is originally $(2n-3) \times 2n$, $(2n-4)$, $(2n-2)$, $(2n-1)$, $2n$ columns and the $(2n-2)$ row will be removed from $R$ to get $\hat{R}$ because they are corresponding to the zero value of $\delta y_{n-2}$, $\delta y_{n-1}$ and $\delta y_{n-2}$. Based on the same reason, $\hat{R}_e$ is derived by removing $(2n-4)$, $(2n-2)$, $(2n-1)$ and $2n$ columns from $R_e$.

Then the set of all the follower nodes is denoted by $V'$ and a subset of the follower node set is $V_f$, $V_f \subset V'$.

$$
\hat{R} = \begin{bmatrix} R(V_f) & X \\ 0 & x_{n-2} - x_{n-1} & x_{n-1} - x_{n-2} & 0 \\ 0 & x_{n-2} - x_n & 0 & x_n - x_{n-2} \end{bmatrix}
$$

$$
\hat{R}_e = (\oplus_{i=1}^{n-3} D_i) \oplus (x_{n-1} - x_{n-2}) \oplus (x_n - x_{n-1})
$$

Here, we suppose that the two outgoing edges from node $i$ are $\{p_i, p_j\}$ and $\{p_i, p_k\}$, so:

$$
D_i = \begin{bmatrix} x_i - x_j & y_i - y_j \\ x_i - x_k & y_i - y_k \end{bmatrix}
$$

4. Stability analysis

In section 3, we propose the formation control laws using the minimally persistent directed graph. In this section the stability of the control laws will be discussed.

The control laws will be expressed by:

$$
\cdot := K \hat{R}_e^{-1} \hat{R} z
$$

where $\hat{R}_e^{-1}$ can be expressed by $\hat{K}_R$ and $\hat{K}_R$ is a diagonal block matrix. Then the above equation is easily transformed into:

$$
\cdot := K \hat{R}_e^{-1} \hat{R} z
$$

Theorem 4.1 For an n-node minimally persistent formation with the agent set $P = \{p_1, p_2, \ldots, p_n\}$ at generic positions, the three co-leaders are indexed by $n$, $n-1$ and $n-2$. $\hat{R}$ is the matrix obtained by removing the $(2n-4)$, $(2n-2)$, $(2n-1)$ and $2n$ columns, and $(2n-2)$ row, from the rigidity matrix $R$. Then a diagonal block matrix $\hat{K}_R$ exists, such that the formation control laws (31) with the minimally persistent are stable.

Proof. In Theorem 3.2 in [17], we obtain the conclusion that if every leading principal minor of $\hat{R}$ is non-zero, then a diagonal $\Lambda$ exists such that the real parts of the eigenvalues of $\Delta \Lambda$ are all negative.

Following Lemma 4.3 and Theorem 4.5 in [17], we know that both $R(V')$ and $R(V_f)$ are generally non-singular ($V_f \subset V'$) with one leader formation.

Finally, for all $1 \leq k \leq n-3$, the outgoing edges from the vertex $k$ are the main factor of $\hat{R}$, so with the leading principal minor of $R(V_f)$, there are three situations to be considered here.

The leading principal minor of $\hat{R}$ is denoted by $D_l (l = 1, 2, 3, \ldots, n-1)$. On account of the sequence appearing in the leading principal minor, only the last appearing node needs to be considered here. Suppose that the last appearing node is $p_k$, and its virtual leader are $p_m, p_n$. The edges from $p_k$ are $\{p_k, p_m\}$ and $\{p_k, p_n\}$.

If $m < k$ and $n < k$, which means that the index of both $p_m$ and $p_n$ are smaller than that of $p_k$ in the sequence of the leading principal minor, then this leading principal minor is $1$ of $R(V_f)$. As analysed in [17], it is known that $R(V_f)$ is non-singular, so the determinant here is nonzero.

1. If $(m < k < n)$ or $(n < k < m)$, for the case of $(m < k < n)$, when both of $p_m, p_n$ are the followers, and as $R(V_f)$ is non-singular, so the determinant is nonzero; when $p_n$ is the co-leader, from Lemma 4.3 and Theorem 4.5 in [17], it is known that the determinant is also nonzero.
2. If $m > k$ and $n > k$, there are three situations that need to be discussed.

When both of them are followers, then it is similar to step (1); when one of them is the follower and the other is a co-leader, then it is similar to step (2); when both of them are co-leaders, then the leading principal minor is:

$$
D_l = \begin{bmatrix} \det \left\{ R(1, 2, 3, \ldots, k-1) \right\} & X \\ 0 & x_k - x_m & y_k - y_m \\ 0 & x_n - x_k & y_k - y_m \end{bmatrix}
$$
where $X$ is a don’t-care vector and $\bar{x}, \bar{y}$ are the $x$-axis and $y$-axis coordinates. So it is also nonzero.

To sum up, the leading principal minor of all the followers is nonzero. Then the leading principal minor of the co-leaders is provided as follows:

$$D_{2n-5} = \text{det} \begin{bmatrix} R(V) & X \\ 0 & X_{n-2} - X_{n-1} & X_{n-1} - X_{n-2} & 0 \\ 0 & X_{n-2} - X_n & X_{n-1} - X_n & 0 & X_n - X_{n-2} \end{bmatrix}$$

$$D_{2n-4} = \text{det} \begin{bmatrix} R(V) & X \\ 0 & X_{n-2} - X_{n-1} & X_{n-1} - X_{n-2} & 0 \\ 0 & X_{n-2} - X_n & X_{n-1} - X_n & 0 & X_n - X_{n-2} \end{bmatrix}$$

Both of them can be transformed into the form as follows:

$$\hat{D} = \text{det} \begin{bmatrix} R(1, 2, 3, ..., n-3) & X \\ 0 & \bar{x} \bar{y} \bar{z} \bar{w} \end{bmatrix}$$

where $\bar{x}$ is the real coefficient gain and $\bar{y}$ indicates the real number, such as $\bar{x}_{n-1}, \bar{x}_{n-2}$ and $\bar{x}_n$. As a result, they are nonzero.

Therefore, the leading principal minor of $\hat{R}$ is nonzero. From Theorem 3.2 in reference [17], we know that a diagonal $\tilde{K}_{c2}$ exists such that the real parts of the eigenvalues of (31) are all negative. **Theorem 4.1** holds.

**5. Simulations**

In this section, simulations using the control laws designed in this paper are presented. Mobilesim software is used to conduct the simulation platforms.

In the simulations, the topology of the mobile robot system is the same as in Fig. 3, in which robots 4, 5 and 6 are defined as the leaders, and the others are the followers.

As analysed in section 3.3, to simplify the analysis, the simulations consist of two stages which are moving along the $x$-axis and then along the $y$-axis. In addition, two kinds of follower topology are studied, which are cyclic and acyclic.

In Fig.5 (a), the topology of the leader set $\{4, 5, 6\}$ is cyclic and the topology of the follower set $\{1, 2, 3\}$ is acyclic.

In the simulations, the gains applied in the controller are defined as follows:

$$\hat{R}_{c}^{-1} \hat{R} = \begin{bmatrix} S(V) & X \\ 0 & S_L \end{bmatrix}$$

where

$$S_L = \begin{bmatrix} -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix}$$

Node $\{1, 2, 3\} \subset V'$, and their topology is acyclic, so based on Theorem 5.2 in [17], we know that all the eigenvalues corresponding to node $\{1, 2, 3\}$ are -1. So only the eigenvalues corresponding to the leader needed to be adjusted by $\hat{K}$.

Snapshots (a) and (b) in Fig. 5 show that the formation is moving with the following initial position:

$$p_1 = (3.8, 1.0), p_2 = (-1.4, 1.8),$$

$$p_3 = (0.5, 5.8), p_4 = (2.2, 2.5),$$

$$p_5 = (3.5, 5.4), p_6 = (5.1, 4.2).$$

Considering the property of $\hat{R}_{c}^{-1} \hat{R}$, here we may choose $k_{n-2}=k_{n-1}=-1$. Fig.5 (c) depicts that the robots have adjusted their orientation along the $y$-axis. In Fig.5 (d), all the robots finally arrive at the destination.

Then, another situation is considered where the topology of the follower set $\{1, 2, 3\}$ is cyclic. The snapshots are shown in Fig 6.

![Simulation snapshots](image-url)
The initial positions of the robots are defined as follows:

\[ p_1 = (-2.5, 5.7), \quad p_2 = (-6.3, 3.8), \]
\[ p_3 = (-2.8, 11.6), \quad p_4 = (1.2, 3.6), \]
\[ p_5 = (2.8, 10.1), \quad p_6 = (5.6, 8.8). \]

Both the leaders and followers are in a cycle, hence it is more complex than the first situation.

From [17], we know that if the topology of \( V'_L = \{ 1, ..., k \} \subset V' \) is cyclic, define \( \alpha \) to be the trace of the cycle weight. Then \( k \) eigenvalues of \( S(V') \) are at \(-1\), and the remaining are at: \(-1 + \alpha e^{\pm \frac{2\pi i}{k}}\), \( \alpha \in (0, ..., k-1) \), where \( \alpha \) is defined as:

\[ \alpha = \frac{(x_{13}y_{35} - x_{35}y_{13})(x_{13}y_{24} - x_{24}y_{13})(x_{23}y_{45} - x_{45}y_{23})(x_{35}y_{24} - x_{24}y_{35})}{(x_{13}y_{14} - x_{14}y_{13})(x_{12}y_{23} - x_{23}y_{12})(x_{21}y_{34} - x_{34}y_{21})(x_{35}y_{24} - x_{24}y_{35})} \]

For this designed persistent formation \( k=3 \), the eigenvalues for \( S(V') \) are:

\[-1, -1, -1, -0.380 \pm 1.073 \, 9i, \]
\[ 4.035 \pm 8.721 \, 2i, 39.888 \, 9 + 70.821 \, 7i \]

\( \hat{K} \) is chosen as follows:

\[
\hat{K} = \begin{bmatrix}
1.1345 & 2.1456 & 0 & 0 & 0 & 0 & 0 \\
-1.0986 & -3.1351 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & -0.0123 & -0.0112 & 0 & 0 & 0 \\
0 & 0 & 10.3246 & 21.6521 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 24.2389 & -18.3465 & 0 & 0 \\
0 & 0 & 0 & 0 & 33.1212 & -19.082 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & -1 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & -1 \\
\end{bmatrix}
\]

The eigenvalues of \( \hat{K}R \) are:

\[ \{ -1.000 \, 0, -1.000 \, 0, \]
\[ -0.283 \, 5 - 0.483 \, 2i, \]
\[ -21.859 \, 2 - 0.387 \, 9i, \]
\[ -2.570 \, 4 + 11.753 \, 0i, \]
\[ -2.570 \, 3 - 11.753 \, 0i, \]
\[ -12.925 \, 2 - 26.539 \, 8i, \]
\[ -0.156 \, 7 + 0.416 \, 3i \} \]

So it can be proved that this situation is also stable.

6. Conclusions

This paper mainly dealt with the problem of minimally persistent formation control with three co-leaders, and the minimally persistent formation control law for the case where three co-leaders are in the cycle was proposed. The non-square matrix, which mostly represents the characteristic of the persistent formation, is successfully utilized to design the control law. In the process of designing the control laws, the straight line motion is substituted by a two-step translation motion along the x-axis and y-axis, respectively. The method of leading principal minor is adopted to prove that the system with the designed law can be stabilized. The simulation results demonstrate that whether or not the followers lie on a cyclic graph, the designed control law is effective.

7. Acknowledgments

The authors would like to thank the Editor-in-Chief, the Associate Editor, and anonymous reviewers for their constructive suggestions to improve the quality of this paper. This work was supported by the Projects of Major International (Regional) Joint Research Program NSFC (61120106010), National Science Fund for Distinguished Young Scholars No. (60925011), NSFC (61175112), and Beijing bionic robot and system the key laboratory.

8. References


